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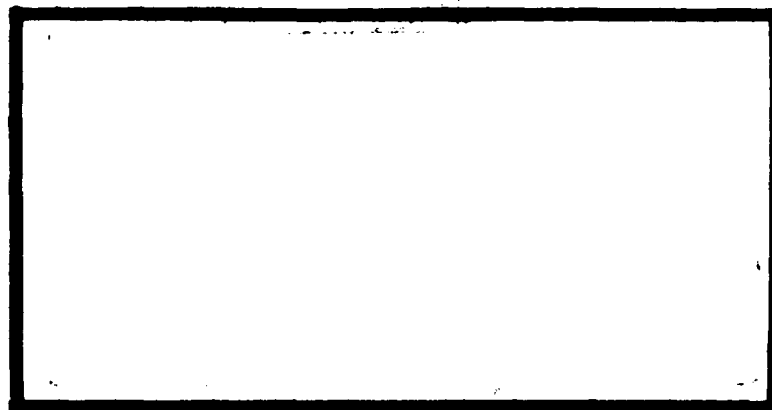
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NOISE CANCELLATION TO IMPROVE
THE QUALITY OF LPC PROCESSED
SPEECH DEGRADED BY NOISE

THESIS

AFIT/GE/EE/81D-10 Christopher L. Batchelor
2Lt USAF

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Presented to the Faculty of the School of Engineering
of the Air Force Institute of Technology

Air University

in Partial Fulfillment of the
Requirements for the Degree of

Master of Science

in Electrical Engineering

by

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Preface

Linear predictive analysis and synthesis of speech is used as a basis for implementing low bit rate voice transmission and minimal digital storage of speech information. The synthesized speech resulting from the linear predictive analysis/synthesis of speech degraded by background noise is of poor quality. This report describes some of the methods proposed to improve the quality of this speech and describes the implementation and performance of one of these methods.

Thanks are due to my thesis advisor, Captain Larry Kizer, for overall guidance through this research. Also, Captain Kizer did a tremendous job ensuring that our Data General computers were maintained and operational. Professor Matthew Kabrisky was most helpful in providing insight into subjective effects of digital signal processing of speech. Lieutenant Robin Simmons is thanked for his assistance in the areas of computer system programming and understanding. Major Ken Castor was helpful in reviewing this report. Lastly, I wish to thank my typist (decoder), Ms. Sharon Gabriel, for doing a good job on this report.

Christopher L. Batchelor

Contents

	<u>Page</u>
Preface-----	ii
List of Figures-----	iv
Abstract-----	v
I. Introduction-----	1
Background-----	1
Problem-----	5
Scope-----	5
Approach-----	5
II. Noise Cancellation Methods-----	7
Time Domain Wiener Filtering-----	7
Frequency Domain Linear Prediction Filter-----	9
Vocoder Filter Bank Analysis-----	11
Linear Prediction Coefficient MAP	
Estimation-----	14
Phase Corrected Spectral Subtraction-----	15
Noise Cancellation Using the Short-Time	
Transform-----	16
III. Implementation of Noise Cancellation Technique--	22
General System Information-----	24
Short-Time Noise Cancellation-----	24
IV. Results-----	28
V. Conclusions-----	36
VI. Recommendations-----	38
Bibliography-----	40
APPENDIX A: Software-----	43
APPENDIX B: The Use of the ILS System-----	65
Vita-----	67

List of Figures

<u>Figure</u>		<u>Page</u>
1	LPC Vocal Tract Model-----	2
2	LPC Vocoder Model-----	4
3	Finite Impulse Response Wiener Filter-----	8
4	Filter Bank Noise Suppression-----	12
5	Old Approach to Spectral Subtraction Noise Cancellation-----	17
6	New Approach-----	18
7	Block Diagram of Short Time Transform Noise Canceller-----	20
8	Block Diagram of Short Time Transform Noise Canceller-----	23
9	Speech Waveform-----	29
10	Speech Plus Noise Waveform-----	30
11	Noise Cancelled Speech-----	31
12	LPC Synthesized Speech Plus Noise-----	33
13	LPC Synthesized Noise Cancelled Speech-----	34
14	Flowchart of Implementation of Noise Canceller-----	44

Abstract

Methods for improving the quality of the speech resulting from linear predictive analysis/synthesis of speech degraded by background noise are discussed. A method of noise cancellation using Wiener filtering in the frequency domain with the short-time Fourier transform was chosen for implementation. Implementation was done on a Data General Nova/Eclipse digital signal processing system in FORTRAN 5. Speech degraded by white gaussian noise was processed through linear predictive analysis/synthesis with and without noise cancellation preprocessing. Preliminary laboratory listenings verified that an improvement in quality was achieved with noise cancellation preprocessing. Although improvement in quality was achieved, more effort is required to make this implementation more efficient and improve the quality of speech produced.

I. INTRODUCTION

Linear predictive coding (LPC) is a successful analysis/synthesis system for bandwidth compression of speech. Research indicates that LPC based systems degrade quickly when processing speech degraded by background noise (Ref 12:6). Thus, it is of interest to apply a digital noise cancellation technique to noise corrupted speech before LPC analysis/synthesis and then evaluate that technique's effectiveness in reducing degradation of quality.

Background

Linear prediction analysis is based on the idea that a speech sample can be approximated as a linear combination of past speech samples. Speech can be modeled as the output of a linear, time-varying system excited by either quasi-periodic pulses (voiced speech) or white noise (unvoiced speech) (Ref 13:38-106). Figure 1 is a block diagram of this model. The time-varying digital filter shown has the steady-state system function of the form:

$$H(z) = \frac{G}{1 - \sum_{k=1}^p a_k z^{-k}} \quad (1)$$

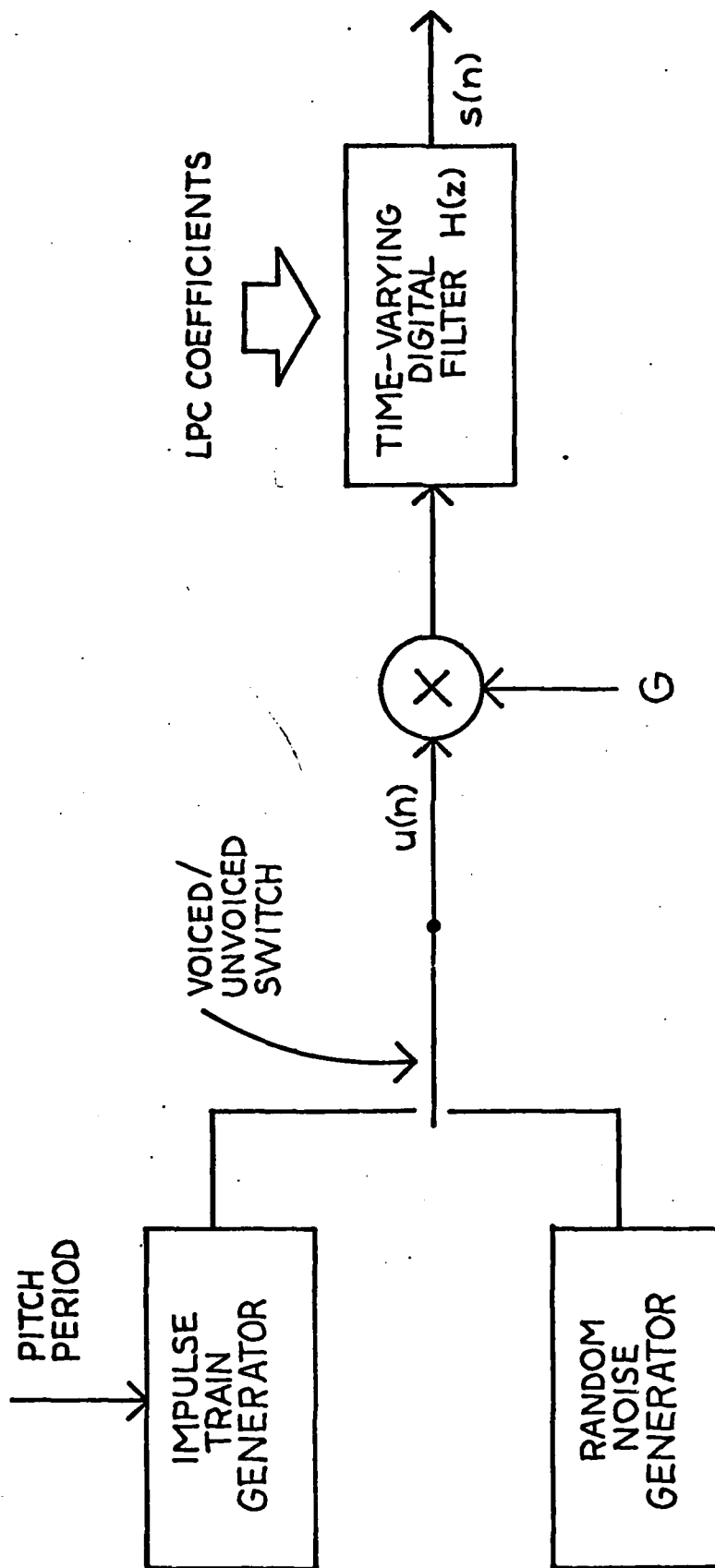


Figure 1. LPC Vocal Tract Model

This system function is an all-pole model and the poles define the resonances or formants of the model as determined by the coefficients $\{a_k\}$. The number p is the order of the model.

The application of linear predictive analysis to encoding speech for low bit rate transmission or storage is termed LPC (linear predictive coding). The LPC analysis parameters are the coefficients $\{a_k\}$, the gain parameter G , the pitch period, and a voiced-unvoiced parameter. Figure 2 shows a block diagram of an LPC vocoder. The transmitter codes the LPC analysis parameters for transmission through the channel and the receiver decodes the parameters and synthesizes the output speech. The LPC analysis parameters can be estimated by many different methods, as described in Digital Processing of Speech Signals by L.R. Schafer and R.W. Rabiner (Ref 13). These methods of estimation for the coefficients $\{a_k\}$ (which determine formants) are not noise tolerant.

Experimental research has demonstrated that four major differences exist between the all-pole linearly predicted spectra of clean and noisy speech (Ref 13:29-30). First, there is a loss of low energy formant information. Secondly, the formant frequencies are shifted. Thirdly, the bandwidth of each formant is wider. Lastly, an overall decrease of spectral dynamic range exists. If the signal to noise ratio is not too low, it has been observed that the primary perceptual effect is generation of "musical tone" like sounds in the background which causes degradation of speech quality

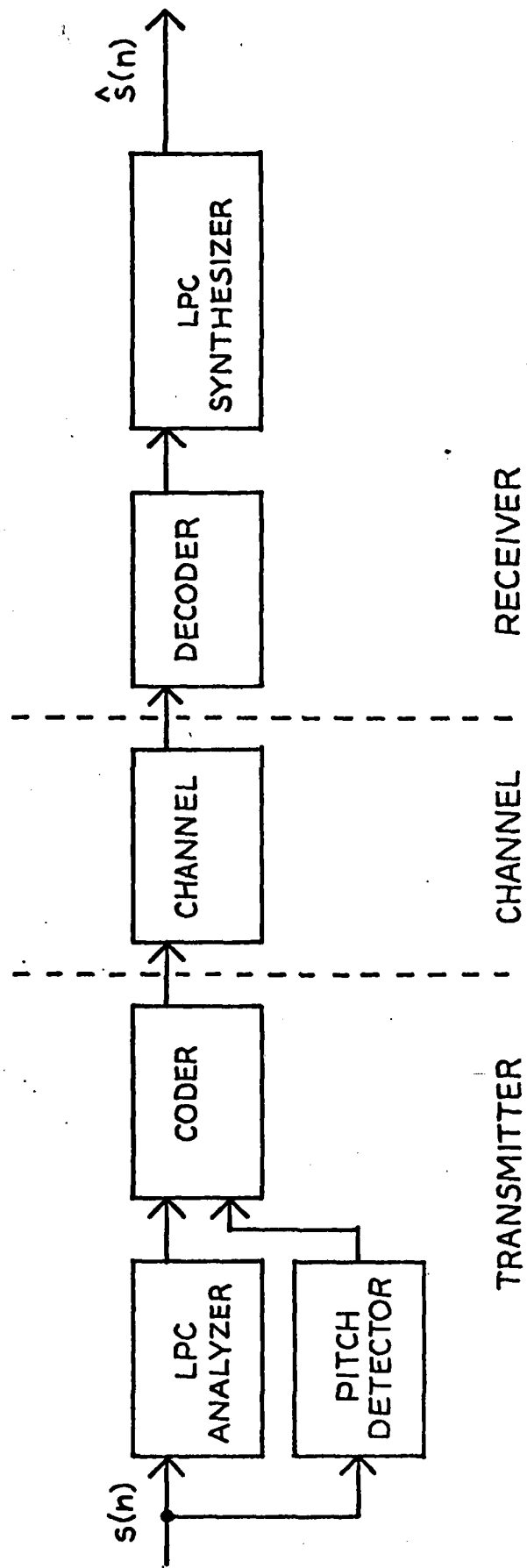


Figure 2. LPC Vocoder Model

(Ref 8:601). Apparently, estimation of the coefficients $\{a_k\}$ is not accurate due to the addition of noise.

Problem

The objective of this thesis is to examine methods to improve the estimation of the LPC coefficients of noisy speech and implement one of these methods on the Data General Nova-Eclipse speech processing system. After implementation, performance of the method was to be determined subjectively.

Scope

The methods to improve estimation of LPC coefficients to be examined will include noise cancellation pre-processing and noise tolerant estimation procedures. This thesis will not discuss estimation of the LPC coefficients using pole-zero modeling because preliminary results indicate that the increased complexity of pole-zero modeling does not improve performance of LPC (Ref 8)

Noise cancellation preprocessing was chosen over noise tolerant estimation procedures for implementation. A noise cancellation preprocessor can be used for other purposes to aid in robust speech processing.

Approach

Each method to improve estimation of coefficients will be described. Also, advantages and disadvantages of each method shall be discussed. Next, a detailed description of

the implementation of noise cancellation technique using short-time Fourier analysis will be given with accompanying results. Finally, recommendations for further research in this area shall be covered.

II. NOISE CANCELLATION METHODS

This section describes techniques which can be applied to improve LPC analysis/synthesis of speech corrupted with noise. The techniques described include time domain Wiener filtering, frequency domain linear prediction filtering, channel noise vocoder filter bank analysis filtering, linear prediction coefficient estimation using the Maximum A Posteriori (MAP) method, phase corrected spectral subtraction, and frequency domain Wiener filtering using the short-time Fourier transform. Unless otherwise specified, the following descriptions are summaries of the reference given.

Time Domain Wiener Filtering

This technique is the application of a Wiener linear prediction filter to reduce additive noise provided that the signal bandwidth is significantly less than the bandwidth of the additive noise (Ref 1). The filter would be applied as a preprocessor before LPC analysis/synthesis.

The implementation is in the time domain using the Widrow-Hoff LMS algorithm. Figure 3 is a schematic diagram of a finite impulse response Wiener filter. The coefficients $W^*(k)$, ($k=0,1,\dots,L-1$) are chosen to minimize the power in the error signal $e(j)$ and satisfy Eq (2). $\phi_{xx}(k)$ is defined as the autocorrelation of the input signal x . $\phi_{xy}(\ell)$ is the cross correlation between the input x and the output y .

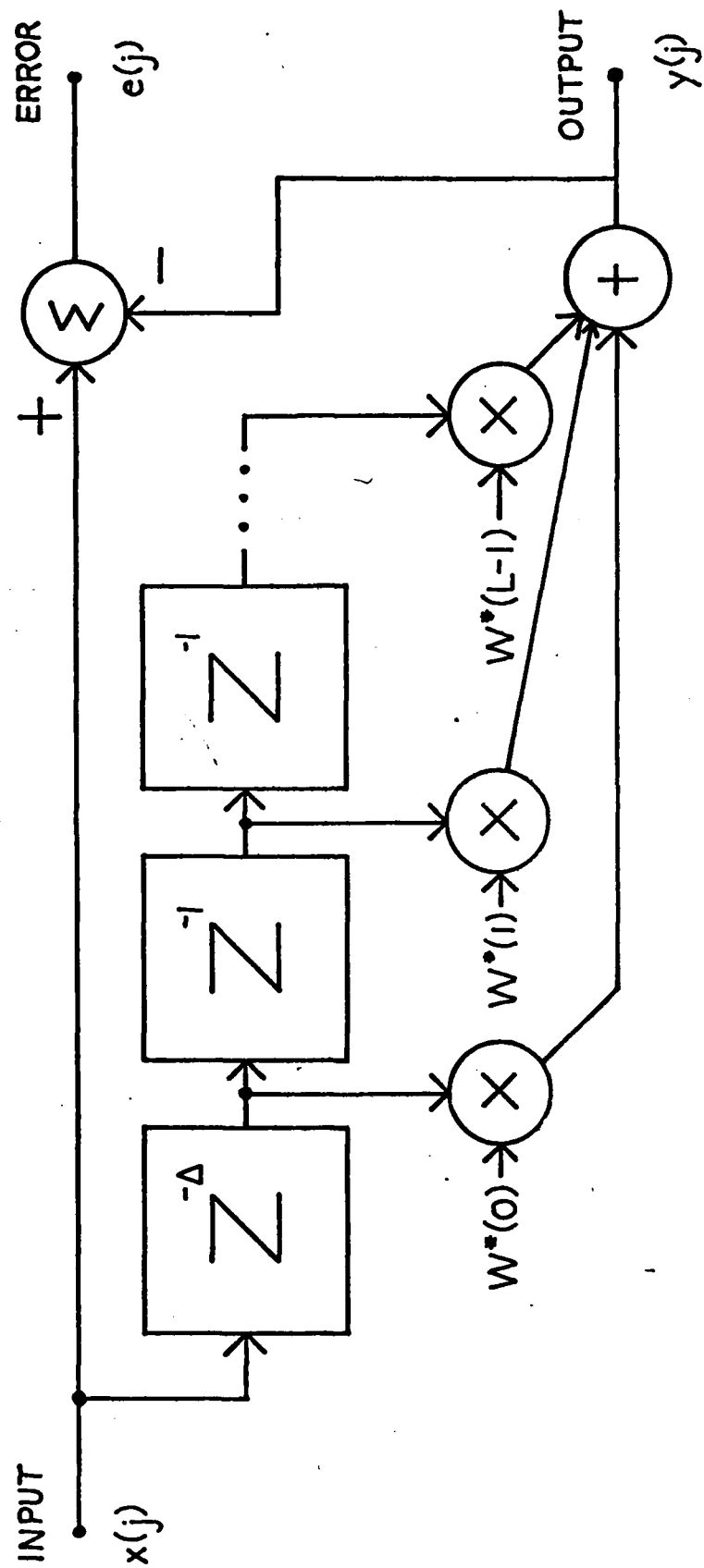


Figure 3. Finite Impulse Response Wiener Filter

$$\sum_{k=0}^{L-1} \phi_{xx}(\ell-k)W^*(k) = \phi_{xy}(\ell+\Delta) \quad (2)$$

Noise suppression results from the fact that the decorrelation time for broadband noise is smaller than that for the narrowband signal. Therefore, it is possible to choose a value for Δ which will prevent the noise components from appearing in the output.

The advantages of this technique are that the prediction distance Δ can be chosen to provide optimum results for the particular noise environment and no external reference noise signal need be provided.

This technique has disadvantages also. First, no subjective experimental results are provided for performance for the filter in conjunction with LPC analysis/synthesis. Research has indicated that echo problems exist with time domain implementations of Wiener filters (Ref 2:694). Lastly, the signal bandwidth must be significantly less than the bandwidth of the additive noise.

Frequency Domain Linear Prediction Filter

This technique attempts to modify the LPC analysis/synthesis process to account for corruption of the speech with noise. Specifically, the speech extraction problem is regarded as a parameter estimation problem (Ref 6).

The method assumes that the power spectral density of the noise is known (noise is stationary) and that the statistics of speech and noise are both gaussian. A periodgram of windowed speech corrupted with noise is calculated by Eq (3).

$$|X_T(f)|^2 = |S_T(f)|^2 + |D_T(f)|^2 + 2 \cdot \text{Re} [S_T(f) \cdot D_T^*(f)] \quad (3)$$

$X_T(f)$, $S_T(f)$ and $D_T(f)$ are the Fourier transforms of windowed noise corrupted speech, speech, and noise, respectively. The unbiased estimate of the speech spectrum is given by Eq (4).

$$|S_T(f)|^2 = |X_T(f)|^2 - E[|D_T(f)|^2 + 2 \text{Re}\{S_T(f) D_T^*(f)\}] \quad (4)$$

This estimate of the speech spectrum is smoothed with a spectral window producing the spectral envelope. Then inverse Fourier transforming the spectral envelope gives the autocorrelation coefficients used in linear predictive analysis (Ref 13:401-403).

The advantage of this method is that noise reduction can be tailored for the particular noise environment encountered by the system.

The disadvantage of this method is that it is not simple to modify existing LPC analysis/synthesis systems to estimate the spectral envelope in this manner.

Vocoder Filter Bank Analysis

This technique performs a spectral decomposition of noisy speech via channel vocoder analysis and attenuates each spectral component depending on how much the measured speech plus noise power exceeds an estimate of the background noise power (Ref 11). This filter would be used as a preprocessor before LPC analysis/synthesis.

A two state model for the speech event is applied in determining the maximum likelihood estimator of the speech power. This model resulted in a class of suppression curves which permits a tradeoff of noise suppression against speech distortion. Real-time experiments have shown that the noise can be made imperceptible by proper choice of a suppression factor, but distortion increases as the input's signal to noise ratio decreases.

The noise suppression filter consists of a bank of second order Butterworth bandpass filters which span the frequency range 120-3270 Hertz. Figure 4 is a block diagram of this system. Measurements must be made to determine the instantaneous signal power and the average signal power at the output of each of the channel filters in order to compute the channel gains. Experimentation showed that a four second

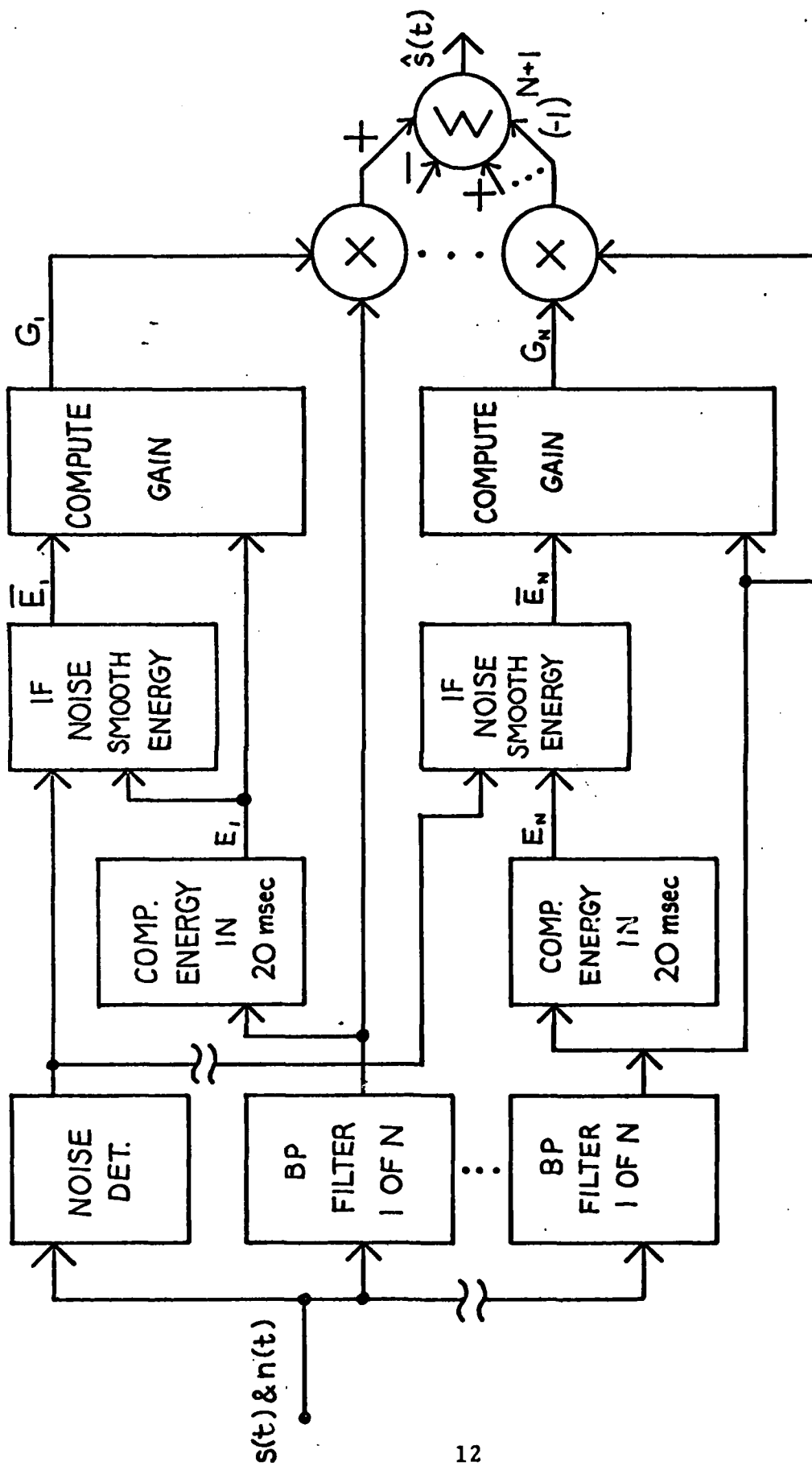


Figure 4. Old Approach to Spectral Subtraction Noise Cancellation

histogram of the frame energies of the signal was bimodal (Ref 11:700). A threshold could be set between the modes, and frames which were speech absent could be determined with high probability. Equation (5) defines the nth channel measurement parameter ($g_n(m)$).

$$g_n(m) = \frac{V_n(m) - V_n(m-1)}{V_n^2(m)} \quad (5)$$

A soft decision point of view determines a class of suppression curves defined by Eq (6).

$$G_n(m) = \frac{1}{2}(1 + \sqrt{g_n(m)}) \frac{\exp(-\xi) I_0(2\sqrt{\xi/(1 - g_n(m))})}{1 + \exp(-\xi) I_0(2\sqrt{\xi/(1 - g_n(m))})} \quad (6)$$

$$\xi = \text{suppression factor} \quad (7)$$

$$I_0(x) = \frac{1}{2\pi} \int_0^{2\pi} \exp(x \cos \theta) d\theta \quad \begin{array}{l} \text{(modified} \\ \text{Bessel function)} \end{array} \quad (8)$$

In a real-time implementation, the measurement parameter $g_n(m)$ is used as a pointer for a table look-up to determine the attenuation $G_n(m)$. To avoid discontinuities, a smoothed gain $\bar{G}_n(m)$ is calculated and applied to the appropriate channel. The channel waveforms are then added together to produce the prefiltered waveform.

This prefilter has three primary advantages. First, it is possible to select a suppression factor which would optimize

intelligibility for a given signal to noise ratio. Secondly, it is possible to integrate the prefilter with efficient channel vocoder implementations. Lastly, no reference noise signal must be provided.

The disadvantage of this method is that it is relatively more complex and would be more difficult to implement.

Linear Prediction Coefficient MAP Estimation

The Maximum A Posteriori (MAP) estimation procedure can be used to estimate the LPC coefficients from speech waveforms degraded by additive white gaussian noise. But this procedure requires solving a set of non-linear equations which require too much computation time. However, the true MAP estimation procedure can be approximated by an iterative method that requires the solution of sets of linear equations (Ref 8).

Equation (9) describes noisy speech $y(n)$ as the sum of speech ($s(n)$) and white gaussian noise ($d(n)$).

$$y(n) = s(n) + d(n) \quad (9)$$

The LPC coefficients can be written in vector form \underline{a} . The MAP estimate of \underline{a} is the vector that maximizes $p(\underline{a}/\underline{y})$ (the probability density of \underline{a} conditioned on \underline{y}). It can be shown that maximizing $p(\underline{a}/\underline{y})$ is a non-linear problem (Ref 8). A "suboptimal" MAP procedure is proposed which estimates \underline{s} and \underline{a} by maximizing $p(\underline{a}, \underline{s}/\underline{y})$. In the iterative procedure, an

initial estimate \hat{a}_0 is obtained, then \underline{s} is estimated by $E\{s/\hat{a}_0, \underline{y}\}$. With estimate \underline{s} , a new estimate \hat{a}_1 is obtained from linear predictive analysis. With the new \hat{a}_1 , the above procedure is repeated obtaining \hat{a}_2 , etc. Estimating \underline{s} by $E\{\underline{s}/\hat{a}_i, \underline{y}\}$ is a linear problem and this iterative procedure converges to a solution that is at least a local maximum of $p(\underline{a}, \underline{s}/\underline{y})$ (Ref 13:201-203). Also, each estimate of \underline{s} by $E\{s(n)/\hat{a}, \underline{y}\}$ can be approximated by filtering $y(n)$ by a non-causal Wiener filter (Eq (10)).

$$H(w) = \frac{P_s(w)}{P_s(w) + P_d(w)} \quad (10)$$

$P_d(w)$ is the power spectral density of the noise.

The primary disadvantage of this technique is that the development is done for the white gaussian noise case. A viable noise tolerant estimation of LPC coefficients must take into account different noise environments.

Phase Corrected Spectral Subtraction

Spectral subtraction noise cancellation preprocessing has been implemented to improve the quality of LPC speech. This method helps, but it introduces "musical noise" problems. A new approach to spectral subtraction noise cancellation has been proposed which avoids the artificial phase distortion which is said to produce the "musical noise" (Ref 10).

Figure 5 is a block diagram of the old approach to spectral subtraction noise cancellation. The time domain output signal is constructed using the phase of the signal plus noise input. Because of this artificial phase distortion, the output does not conform to a linear predictive model. According to the new approach, this problem is eliminated when the subtraction is done at the point where the autocorrelation is calculated or when the covariance matrix is computed in linear prediction analysis.

Figure 6 is a diagram of the new approach in which the autocorrelation lags are modified. If the input noise is assumed white, then the pre-whitening filter can be discarded.

The covariance matrix can be modified also to effect noise cancellation. For white noise, only the diagonal terms of the covariance matrix need be reduced by the noise power for the matrix to conform to the noise-free case. The non-white noise case would require a time varying pre-whitening filter.

The primary advantage of this method is that no noise reference channel need be provided.

Noise Cancellation Using the Short Time Transform

The time domain implementation of a Wiener least squares filter to perform noise cancellation is ineffective when the noise characteristics, e.g., mean, variance, etc. change

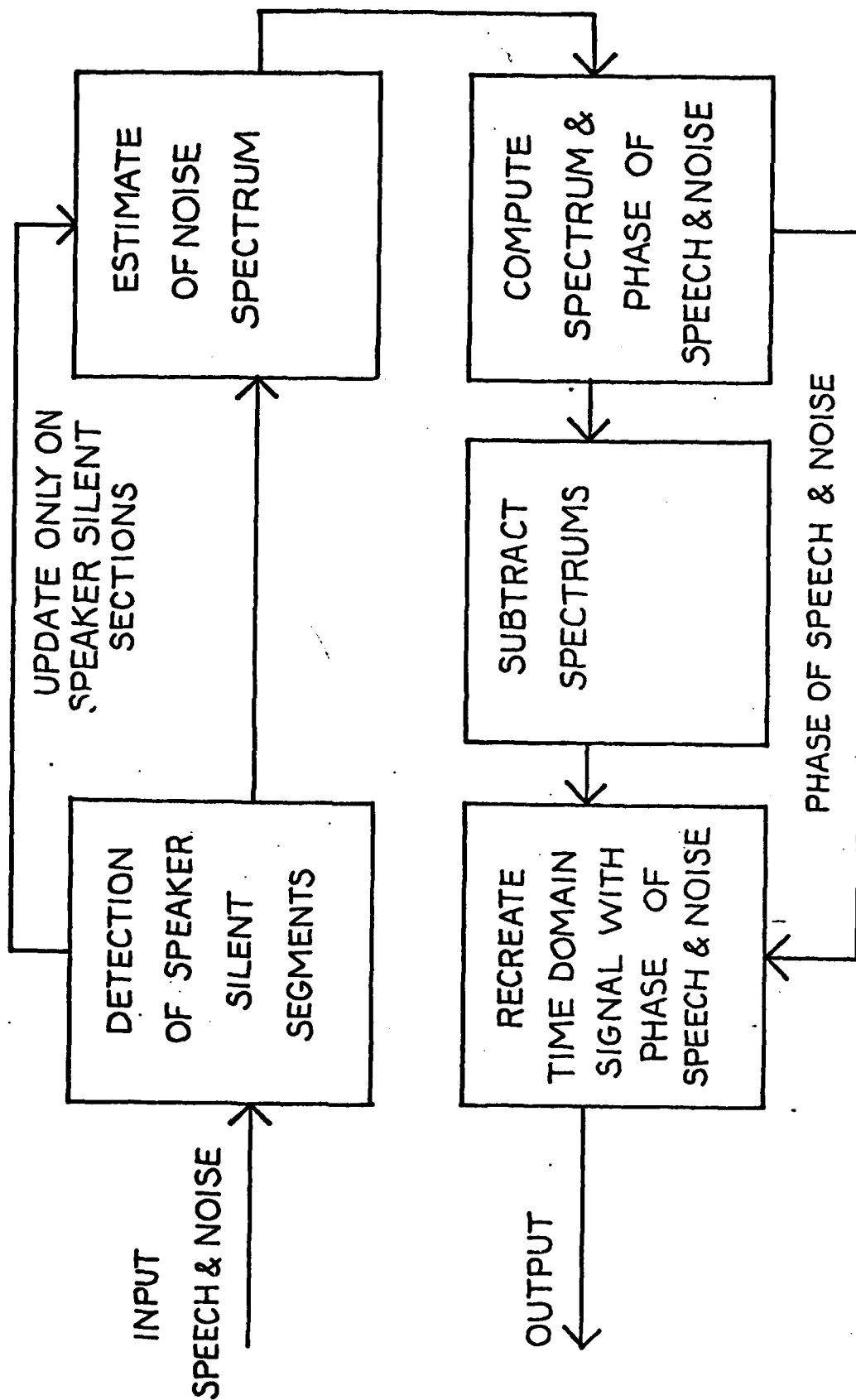


Figure 5. Filter Bank Noise Suppression

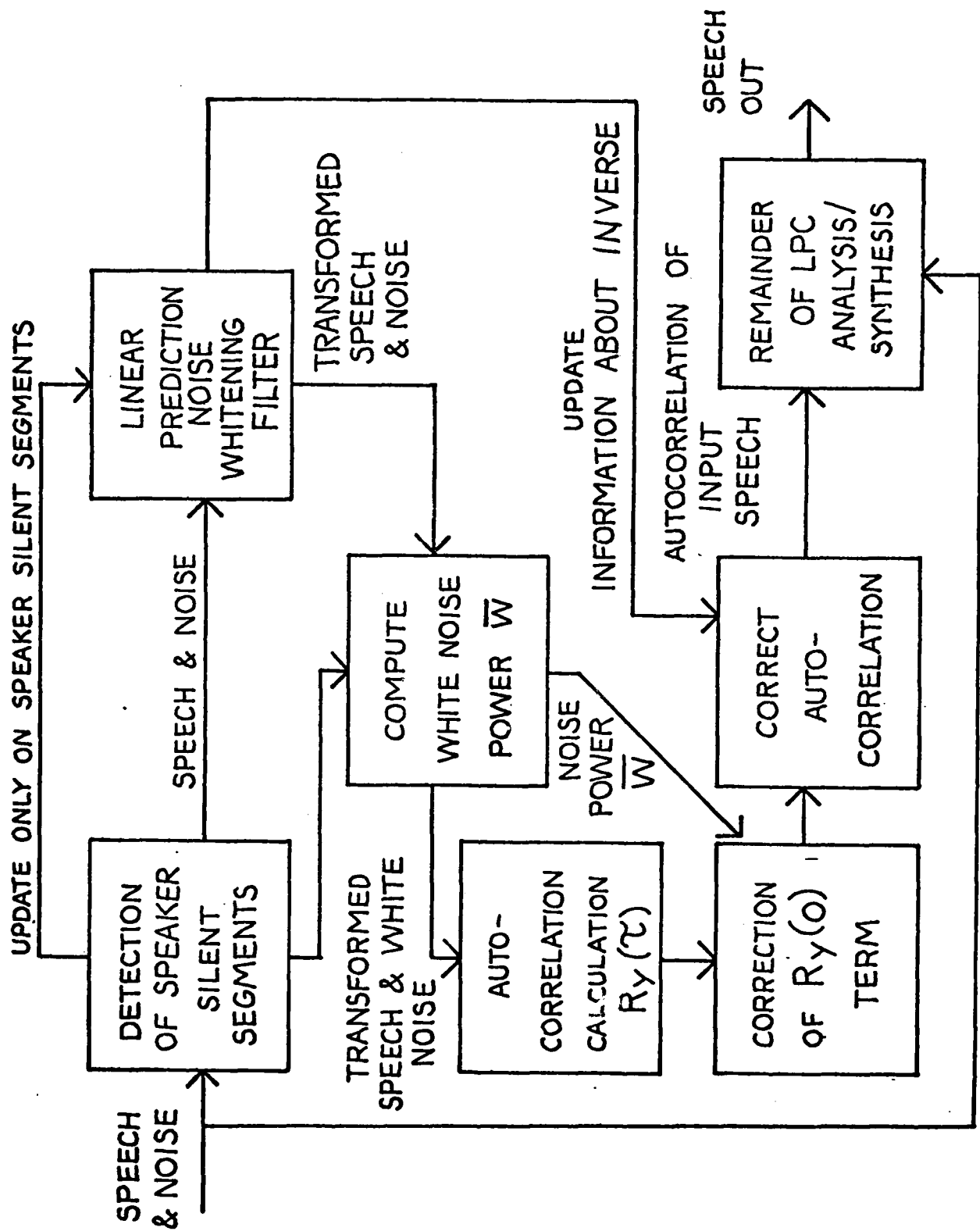


Figure 6. New Approach

rapidly with time. Thus, a frequency domain approach using the short time Fourier transform (FFT) to estimate the required Wiener filter is proposed (Ref 2). Using the efficiency of the FFT results in a computation rate which is proportional to the filter length times the log of the filter length. Therefore, the FFT approach is a viable alternative for real time implementation.

Figure 7 is a block diagram of the filter's construction. The signal x represents speech plus noise and v represents a noise reference channel. $S_{vx}(f)$ is the cross power spectral density between speech plus noise and noise. $S_{vv}(f)$ is the power spectral density of the noise. $W(f)$ is the estimated Wiener filter. A more complete description of this filter will be given in the next section, because this filter was chosen for implementation.

There are three primary advantages of this approach. First, experimental comparisons have shown that the computational efficiency is 3.5 times as great as time domain methods. This means that, for reverberant high noise environments requiring large filter lengths, implementation in real time may be accomplished. Secondly, although the filter structure shown in Figure 7 requires a reference noise channel, it is possible to obtain estimates of the background noise spectrum during speaker-silent segments. Lastly, no a priori information about the energy in the reference channel is needed in order to pick a prediction

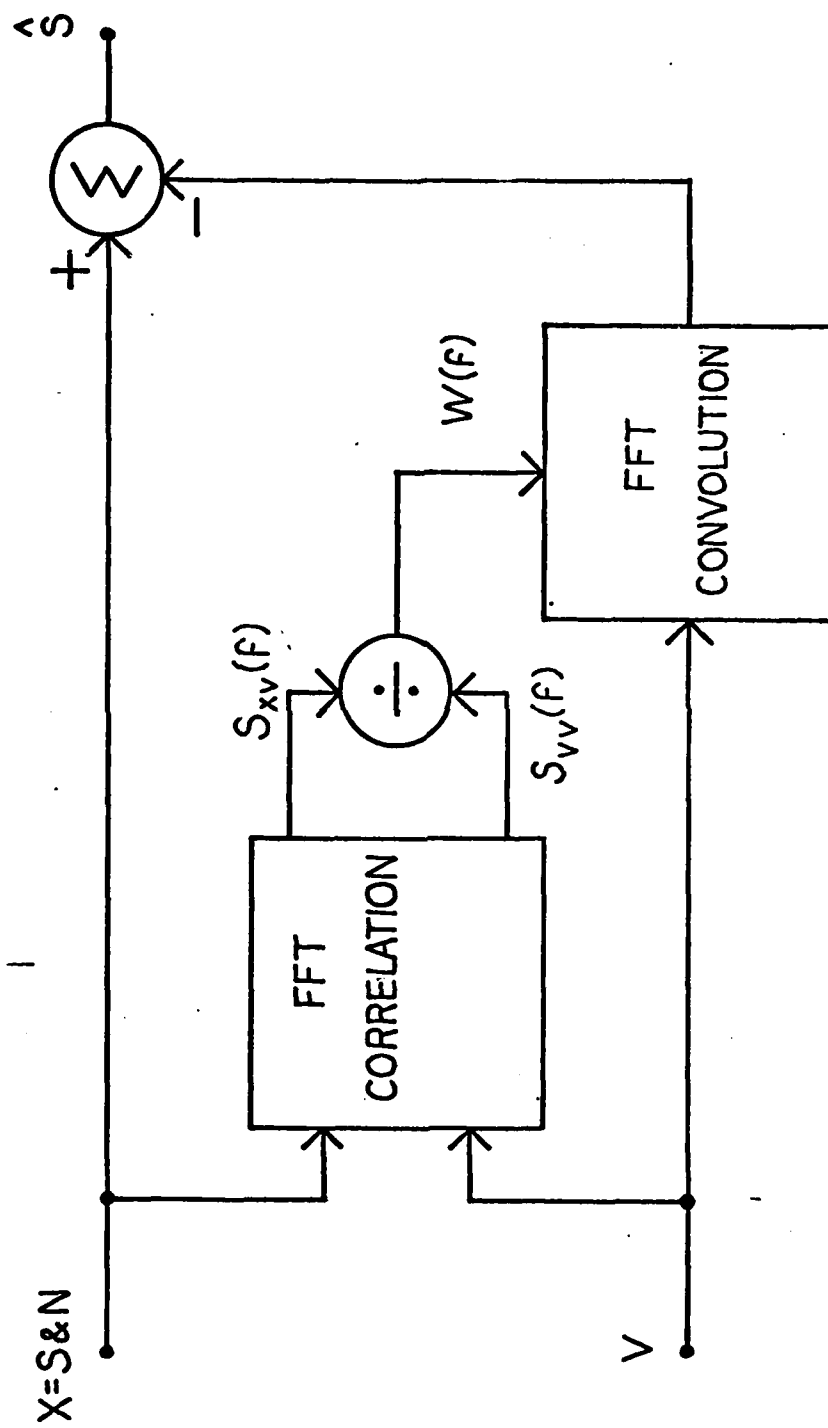


Figure 7. Block Diagram of Short Time Transform Noise Canceller

distance as in some time domain methods. If the prediction distance is marginally too large, then the algorithm will produce echo. If too small, the algorithm will be slow to converge. This gain adjustment is taken care of automatically in the frequency domain approach. Therefore, higher quality output is produced free from echo which has been demonstrated already.

There are two disadvantages of this method. It will require considerably more memory than time domain methods. Also, it will be much more complex to program than time domain methods.

III. Implementation of Noise Cancellation Technique

I chose to implement the short-time transform technique of noise cancellation. This technique was described in the previous section and is depicted in Figure 8. This technique was chosen for the following reasons. First, it is a preprocessor and thus can be used directly to improve LPC processing without modification of the LPC implementation. Also, as a preprocessor, it may be used to enhance the performance of other speech processing procedures being developed (one example is phoneme recognition). Secondly, this frequency domain technique had faster adaptation time than time domain implementations of the Wiener filter (Ref 2). Thirdly, preliminary subjectivity results indicated that frequency domain methods do not have echo problems, as discussed in the last section. Lastly, it could be implemented in the time allowed with greater ease than the frequency domain vocoder method discussed earlier. The frequency domain vocoder shares with the short-time transform method the advantages of faster adaptation time and lack of echo problems, but is much more complex and more difficult to implement. Therefore, the short-time transform method was chosen over methods which modified the LPC implementation, were implemented in the time domain, and required too much complexity for implementation in the time allotted.

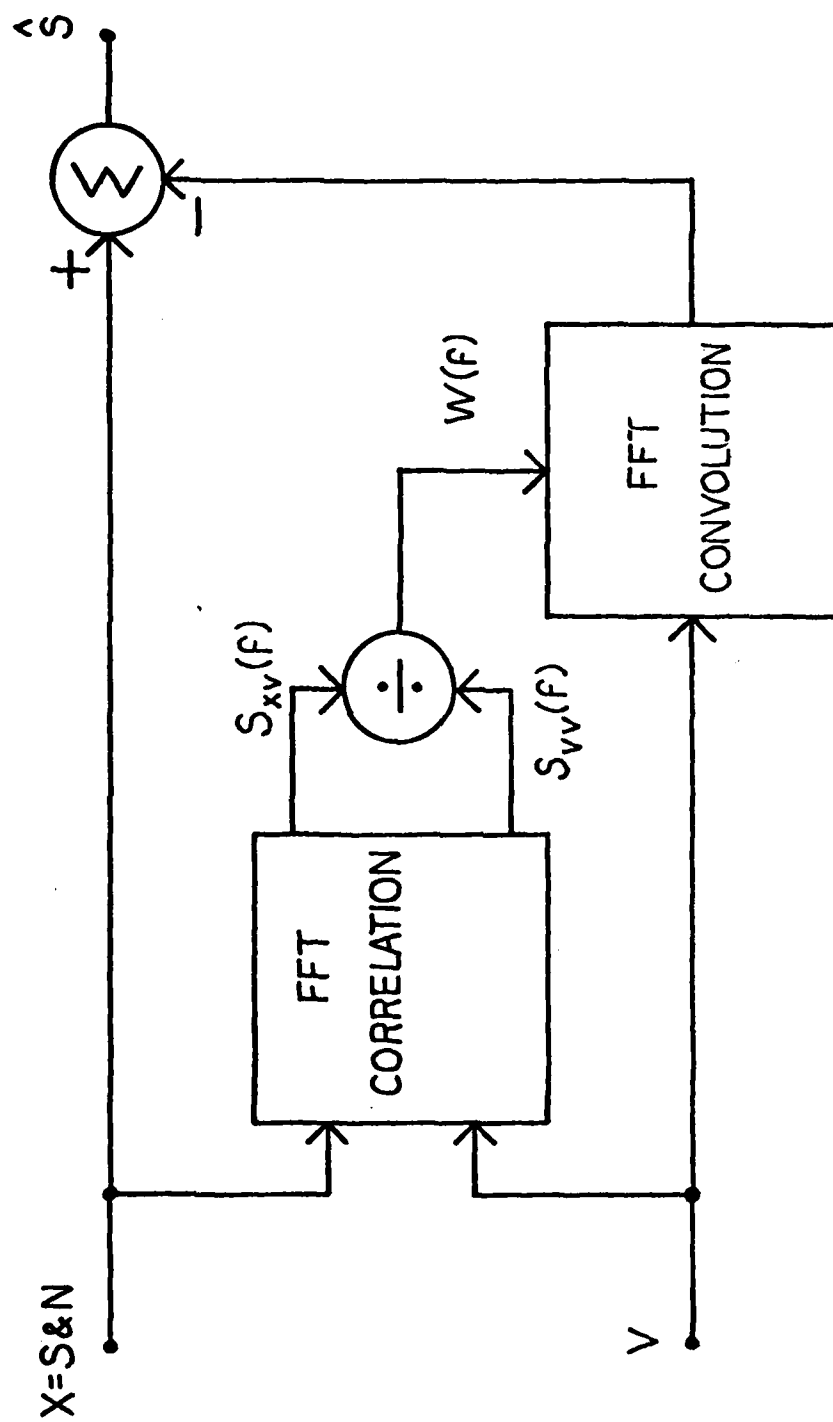


Figure 8. Block Diagram of Short-Time Transform Noise Canceller

General System Information

This technique was implemented in Fortran 5 on a Data General Nova-Eclipse signal processing system. Analog speech data are low-pass filtered at 4.0 kilohertz and sampled at 8000 samples/second (approximately the Nyquist rate). Speech files consist of 88 blocks (256 words/block) of integers ranging from -2048 to 2048 (-5 to +5 Volts). More specific information about input/output of speech files through program AUDIOHIST, written by Paul Finkes, is available (Ref 14).

Short-Time Transform Noise Cancellation

The implementation of this frequency domain method is based on an article by Lawrence R. Rabiner and Jant B. Allen, titled "Short-Time Fourier Analysis Techniques for FIR System Identification and Power Spectrum Estimation" (Ref 14). Specifically, Rabiner and Allen define the short-time Fourier transform of a signal $x(n)$, at time mR as Eq (12):

$$X(m,k) = \sum_{n=mR-L+1}^{mR} x(n)w(mR-n) \exp(-j \frac{2}{N} kn) \quad (12)$$

R is the period between estimates of the short-time transform of the signal and $w(n)$ is a causal FIR window of duration L samples. They show if the signal plus noise signal input is defined as $x(n)$ and the noise input $v(n)$, then the unbiased Wiener filter estimate is given by Eq (13).

$$H(K) = \frac{S_{vx}(K)}{S_{vv}(K)} \quad (13)$$

where

$$S_{vx}(K) = \sum_{m=0}^{p-1} \sum_{q=q_1}^{q_2} X(m, K) V^*(m+q, K) \quad (14)$$

$$S_{vv}(K) = \sum_{m=0}^{p-1} \sum_{q=q_1}^{q_2} V(m, K) V^*(m+q, K) \quad (15)$$

and

$$q_1 = \text{integer part of } [(L+\hat{M}-2)/R]$$

$$q_2 = \text{integer part of } [(L-1)/R]$$

$$\hat{M} = \text{estimate of the system's impulse response duration}$$

$$p = \text{number of analysis sections.}$$

Bounds are defined for the parameters L and R in Eqs (16) and (17).

$$L \geq \hat{M} \quad (16)$$

For a Hamming window,

$$R \leq L/4 \quad (17)$$

For this implementation, \hat{M} was chosen to be equal to L or 128 samples (16 msec). R is chosen to be 32 samples (4 msec). The parameters q_1 and q_2 are then calculated as -7 and 3. The number of analysis sections p is left as a variable whose effect is to be evaluated because no information relating to this specific application was available in order to choose a value. Rabiner and Allen do show that, as p increases, less error is made in the estimate $H(K)$ (Ref 14: 190-191). The parameter p is related to the total number of points used in the analysis N' by Eq (18)

$$p = \text{integer part } \left[\frac{(N' - L + R)}{R} \right] \quad (18)$$

The process of noise cancellation is implemented can be described as follows. First, weight 128 point sequences of both channels with Hamming windows and compute N point FFT by zero filling from 129 to N. Repeat previous step (except shift 32 samples) until $p + q_2 - q_1$ FFTs have been calculated and stored. Then, calculate $S_{vx}(K)$ and $S_{vv}(K)$ according to Eqs (14) and (15) by correcting the relative phase between successive FFTs at each analysis frame M due to the shift of 32 samples between successive FFTs. $H(K)$ is then calculated according to Eq (13). A FFT of the noise reference channel $V(0,K)$ is multiplied by $H(K)$ and then inverse Fourier transformed to produce the sampled sequence of correction signal

on the interval 1 to 128. The stored FFTs of the following two channels are updated by shifting in memory and storing one new FFT of both channels in the space opened up after shifting. Then $H(K)$ is calculated again as above, repeating the process. A new sampled sequence of the correction signal is produced and added with a 75% overlap of the previous sequence calculated. After reconstruction, the correction sequence is subtracted from the speech plus noise channel, completing the operation. Appendix A gives a description of how this process was implemented.

IV. Results

The short-time noise cancellation implementation was used to process files consisting of 2.0 seconds of speech plus noise. Each speech plus noise file was generated with a predefined speech to noise ratio by program SPLUSN. The noise used as the reference channel and by SPLUSN is white gaussian and generated by program NOISE. Both NOISE and SPLUSN are described in Appendix A.

The time required to process each 2.0 second speech file using six analysis frames per estimation was excessive (approximately 5.2 hours). The execution time is approximately proportional to the number of frames per estimation since processing with three analysis frames per estimation required 2.6 hours. The total run-time can be approximately broken down into three areas. It took 27.8% of the time to read and write complex arrays to memory. It took 4.22% of the time to calculate 1024 point DFTs. Also, it took 72.16% of the time to perform all other processing which mostly included complex arithmetic and logical operations.

The noise cancellor performance is depicted by Figures 9, 10 and 11. Figure 9 is a plot of four sequential segments (each segment is .128 seconds long) of speech. Figure 10 is a plot of the same speech segments plus noise (SNR = 0 db, as defined by program SPLUSN in Appendix A using B weighting curves on speech and noise energies). Figure 11 is a plot of the output of the noise cancellor when the speech plus noise

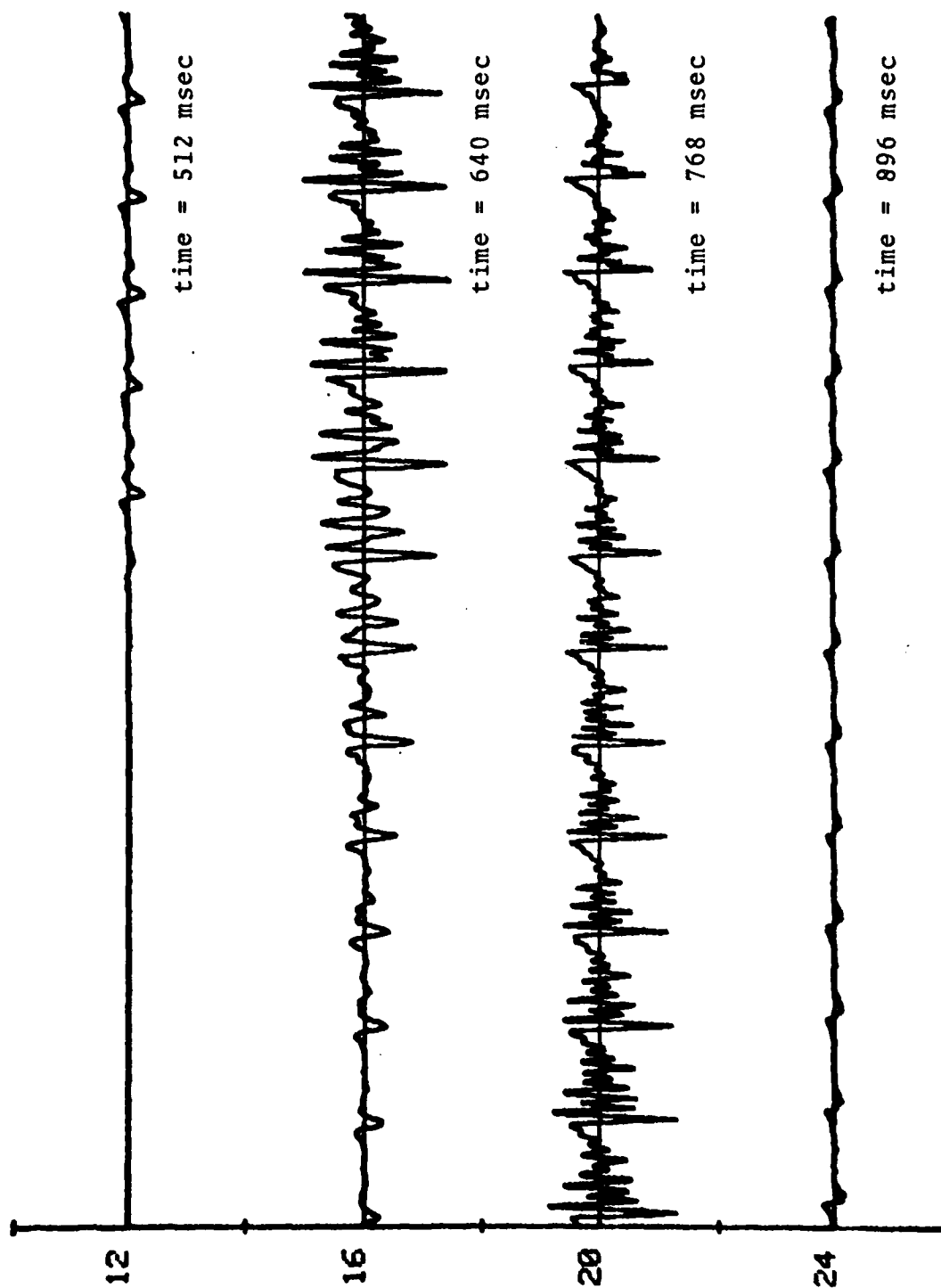


Figure 9. Speech Waveform

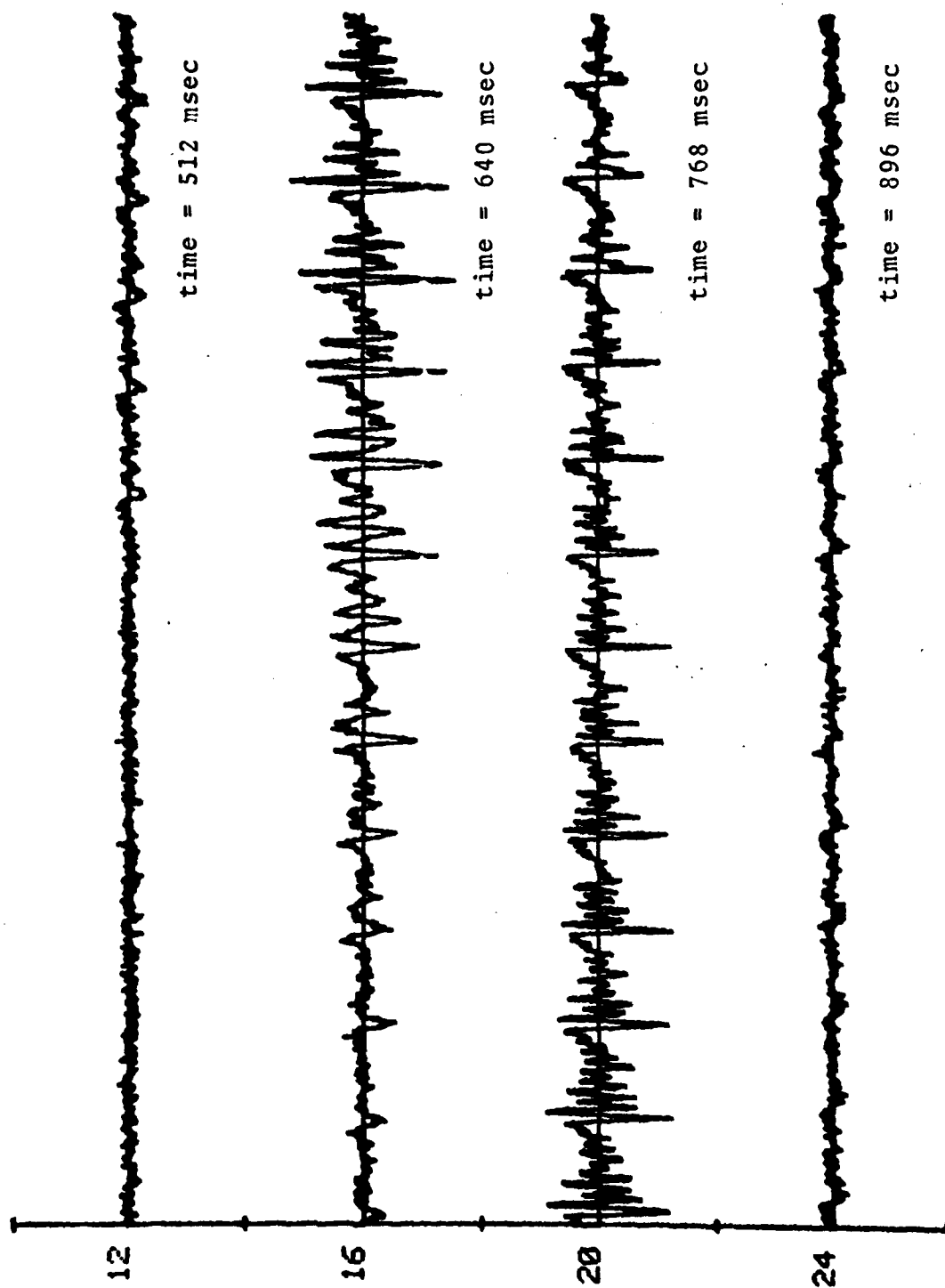


Figure 10. Speech Plus Noise Waveform

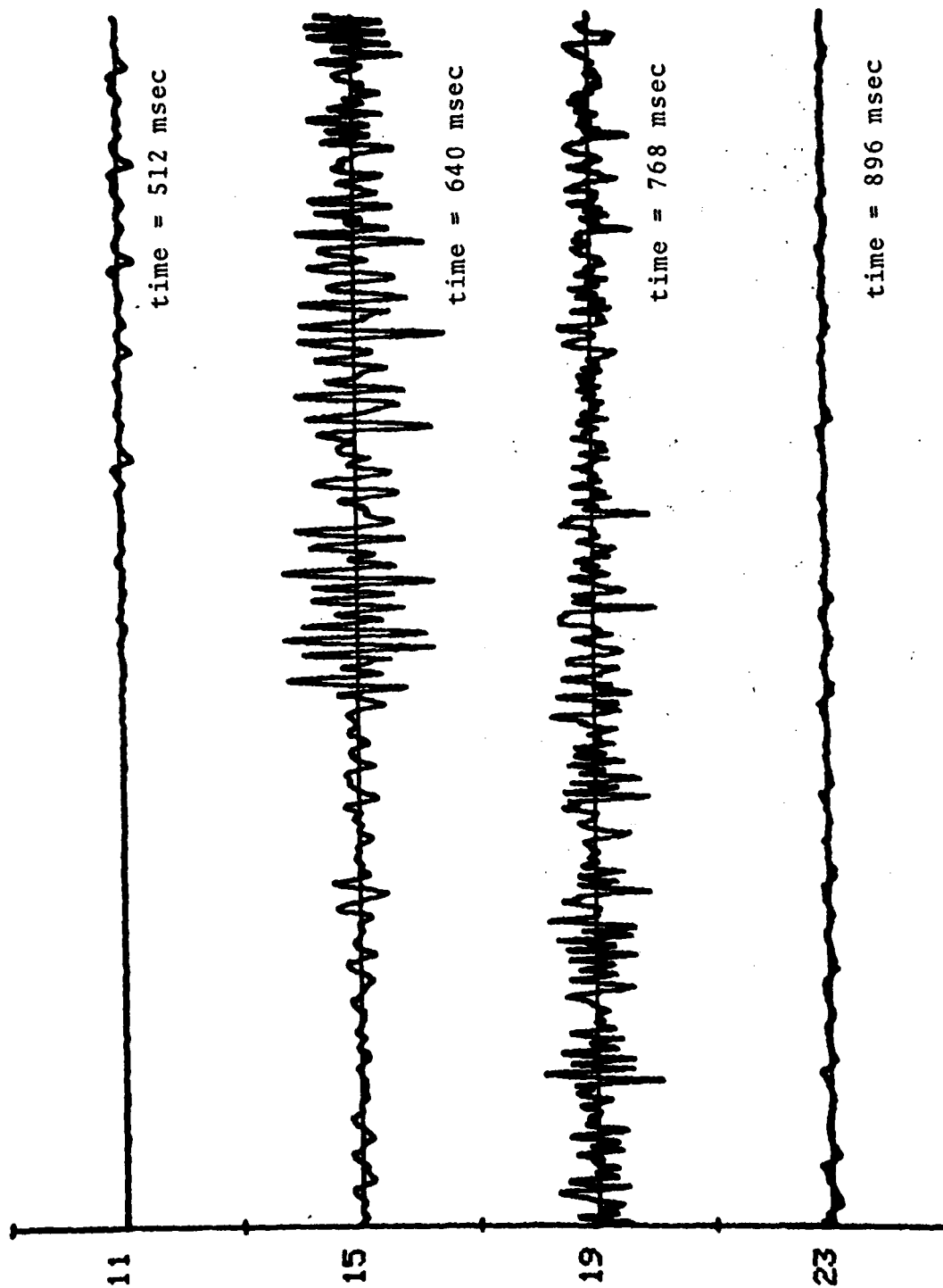


Figure 11. Noise Cancelled Speech

in Figure 10 is used as its input and six analysis frames were used per estimation. Note that little noise appears in the output, but the speech waveform is distorted. It is more difficult to visually pick out the glottal pulse in the speech waveform. I listened to output speech and it sounded like whispered speech. Also, little noise was heard in the output.

Noise cancellation was performed with three and six analysis frames per estimation.

It was noted that the output speech was less whisper-like (higher quality) with six analysis frames per estimation. However, the output speech was still whisper-like and of low quality.

Next, I used our Interactive Laboratory System (ILS) to perform linear predictive analysis/synthesis on speech plus noise and noise cancelled speech. The use of the ILS system in this application is described in Appendix B. Also, the details on how the ILS system performs linear predictive analysis/synthesis are given in the ILS application note number 1 entitled "Speech Analysis and Synthesis" (Ref 15).

The number of points per analysis frame used in the LPC analysis was chosen to be 128. The number of coefficients estimated was chosen to be 10. Also, analysis frames were windowed with a standard Hamming window.

The synthesized speech is depicted in Figures 12 and 13. Figure 12 is the LPC processed version of the speech plus noise shown in Figure 10. Figure 13 is the LPC processed

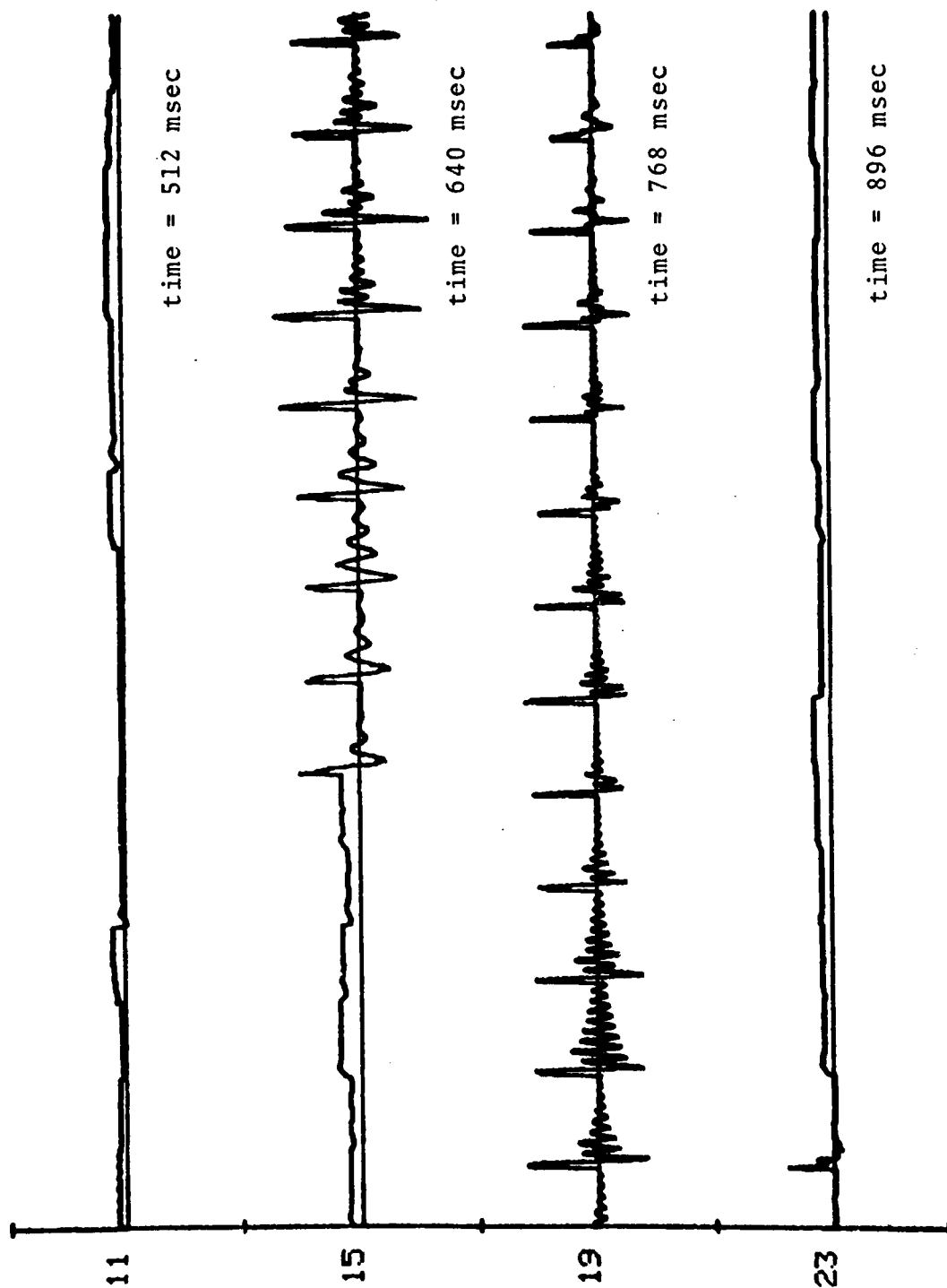


Figure 12. LPC Synthesized Speech Plus Noise

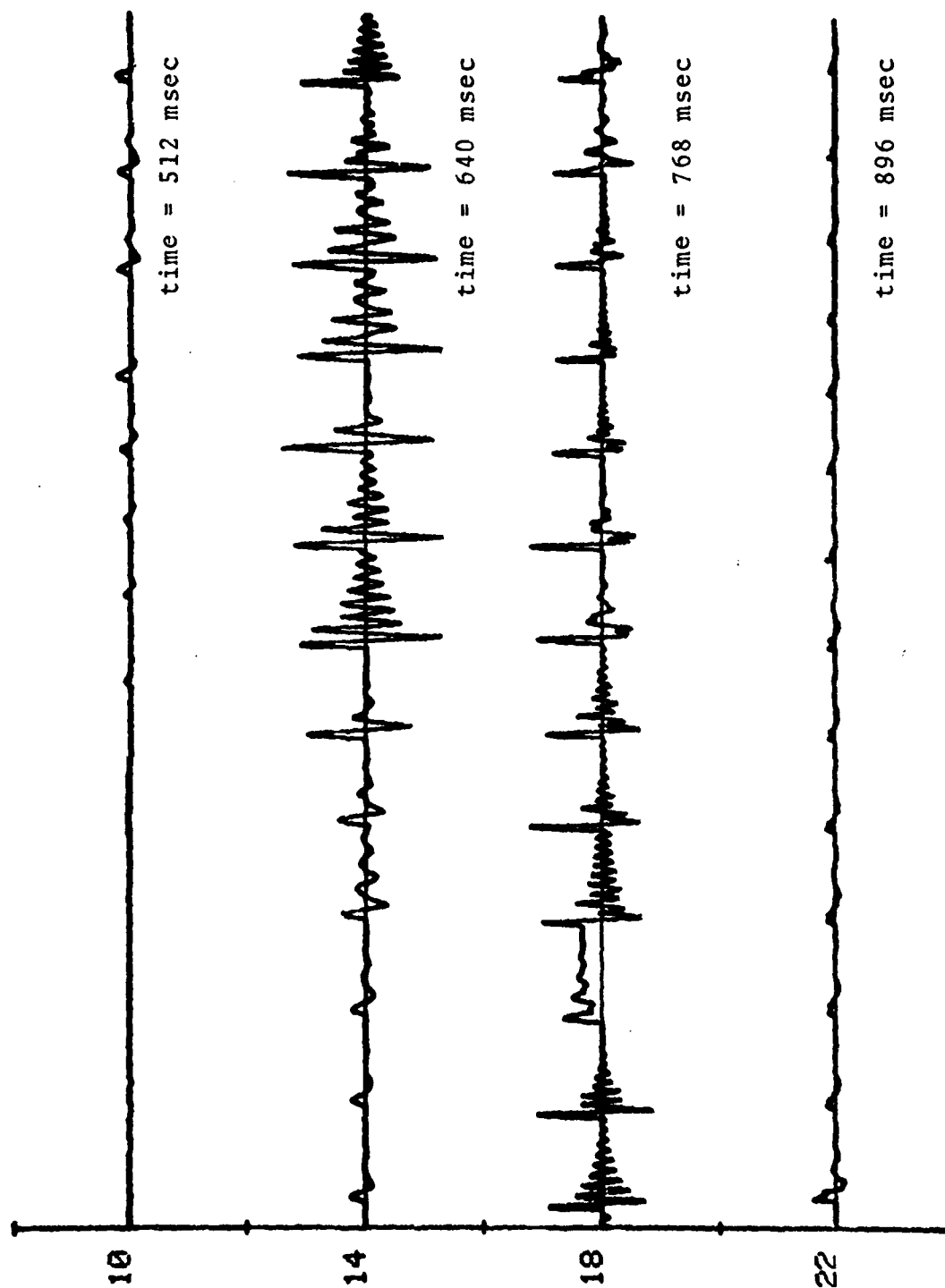


Figure 13. LPC Synthesized Noise Cancelled Speech

version of the noise cancelled speech shown in Figure 11. After listening, the LPC processed speech plus noise was perceived to be of very poor quality. The LPC processed noise cancelled speech retained some of its whisper-like quality, but was of much higher quality than the LPC processed speech plus noise.

V. Conclusions

This short-time noise cancellation method was capable of improving the quality of LPC processed speech plus noise, but this implementation has two major weaknesses. First, this implementation takes an excessive amount of time to perform noise cancellation. Secondly, the output speech produced by this implementation sounded like whispered speech.

This implementation takes an excessive amount of time for the following three reasons. First, 27.8% of the processing time is devoted to reading and writing 1024 point complex arrays (4096 words) to disk. Secondly, the Data General Nova/Eclipse performs floating point instead of integer complex arithmetic. Integer arithmetic is carried out quicker, but floating point arithmetic allows greater dynamic range in processing. Lastly, the DFT size was chosen to be 1024 and may be reduced to 512 without reducing the performance of the cancellor. A DFT size of 512 would cut in half the time necessary to read and write complex arrays to disk and the time required for complex arithmetic operations.

The whispered speech effect of the output from the cancellor was reduced when the number of analysis frames used per estimation was doubled from three to six. The effect of increasing the analysis frames used in each calculation of estimates of the cross-spectrum and the auto-

spectrum is that these spectra are smoothed more over time. Thus, the Wiener filter estimate does not change as rapidly over time and will not produce as much modulation of the speech waveform. Additional smoothing could be implemented by smoothing the spectrum estimates by taking partial sums of past and present spectrum estimates.

VI. Recommendations

The areas for future research into the characteristics of performance of this implementation of the short-time transform noise cancellation include the following:

1. Whether or not the current implementation is modified, formal subjective testing (diagnostic rhyme test) should be carried out with LPC processed speech plus noise and LPC processed noise cancelled speech. This testing would provide a more firm basis for describing the performance of the noise cancellor.
2. The addition of an array processor on the Data General Eclipse computer could greatly increase the speed of complex array processing. This noise cancellor would have to be modified to allow the array processor to perform arithmetic operations on complex arrays.
3. The addition of more memory allocated for programs may make it feasible to store FFTs in extended memory instead of on disk. Thus, the program would not have to perform input/output operations to disk which take 27.8% of the current processing time.
4. This noise cancellor's FFT size of 1024 points could be halved to 512 points and cut in half the processing time. But, it must be determined whether

or not the decreased resolution in frequency reduces the noise cancellor's ability to reject single interfering tones.

5. Spectral smoothing of the power spectrum estimates would probably improve the quality of the noise cancelled speech. This could be accomplished by taking partial sums of past and present power spectrum estimates.
6. The current implementation requires an external noise reference channel. This implementation could be modified to update the estimate of the power spectrum of the noise on speaker silent segments from the speech plus noise channel. This requires detection of the speaker silent segments by thresholding. The threshold is chosen by studying four second histograms of speech plus noise which are bimodal. A threshold is picked from between the two modes.

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APPENDIX A

Software

Noise Cancellation Process

The noise cancellation process consists of three programs executed in sequence. First, BEGNCL obtains names of I/O files and number of analysis frames per estimation from user and stores them for the following programs to read. Second, NCANCEL performs the calculations necessary to produce the (correction signal) time domain estimate of the noise. Thirdly, SUBTRACT subtracts the time domain estimate of the noise from the speech plus noise file. This entire process is executed by running macrofile NC.MC. Figure 14 is a flowchart of the overall process. The following source listings explain each program.

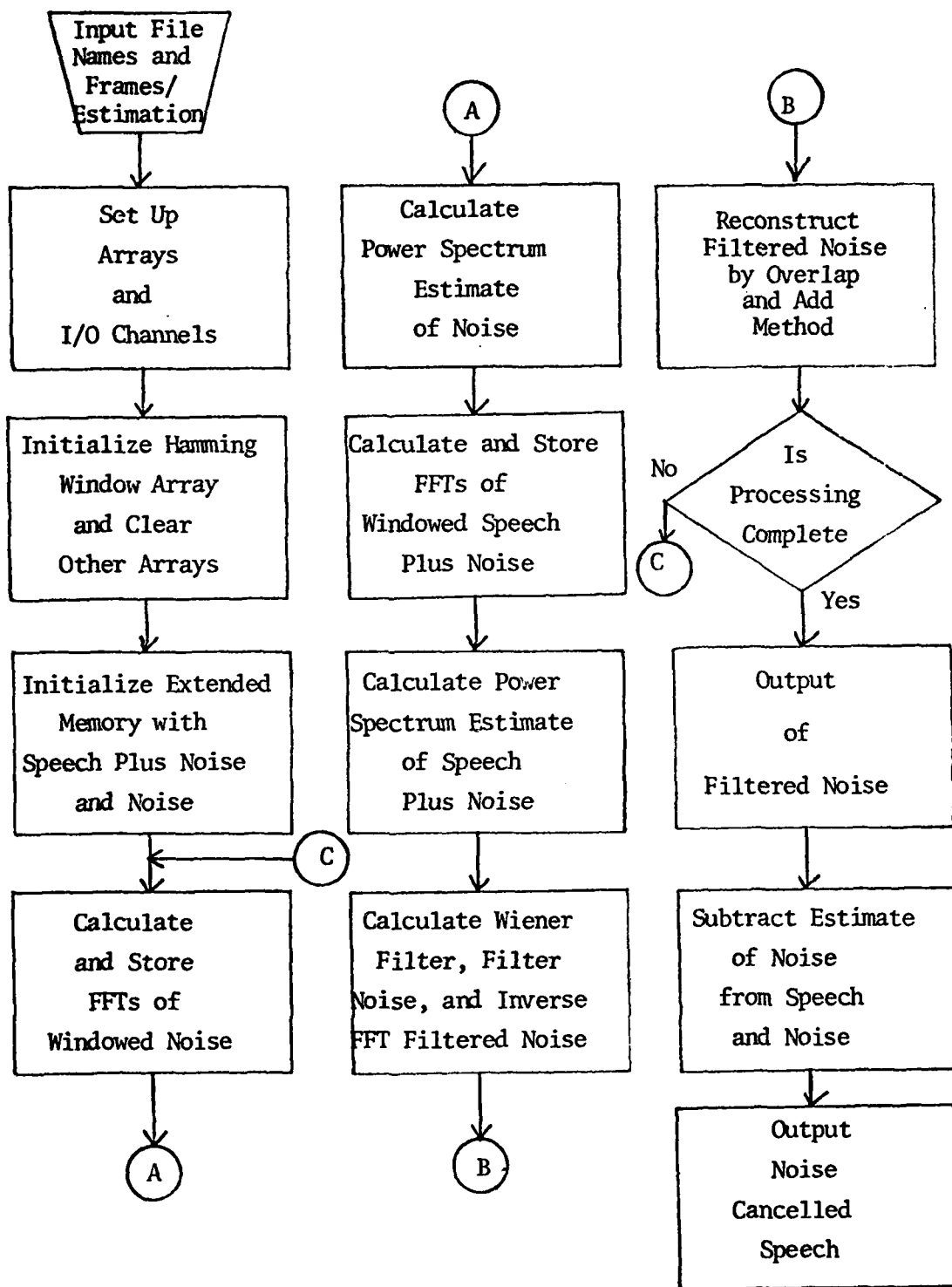


Figure 14. Flowchart of Implementation of Noise Canceller

U U

U U

[illegible]

U U

U U

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[illegible]

U U

(The above row contains 28 uppercase 'U' characters.)

U U

U U

U U


```

EQUIVALENCE(FOURN(1),I02(1))
OPEN 5,"NSFILT",ATT="BC",REC=62 ;open correction signal file
OPEN 3,"NAME" ;open file containing names of I/O files
READ(3,100)NSPFIL(1) ;get name of speech plus noise file
READ(3,100)NOISFIL(1) ;get name of noise file
READ(3,100)NGOUTFIL(1) ;get name of output file
READ(3,101)NANAL ;read # of analysis frames/estimation
FORMAT(S13)
100 FORMAT(I4)
101
OPEN 2,NOISFIL ;open noise file
OPEN 4,NSPFIL ;open speech plus noise file
NUMTF=10+NANAL ;calculate # of FFTs stored for calculation
TYPE" NUMBER OF ANALYSIS FRAMES/ESTIMATION=" ,NANAL
OPEN 1,"PSPEC",ATT="BC",REC=((3*NANAL+12)*16) ;open FFT storage file

C**** INITIALIZE ARRAYS
C
DO 1 I=1,128 ;initialize Hamming window array
  HAMMING(I)=CMPLX(.54-.46*COS(2*3.14159*(I-1)/127.0),0.0)
DO 2 I=1,1024 ;clear the following arrays
  SPEC(I)=CMPLX(0.0,0.0)
  FOURN(I)=SPEC(I)
  FOURS(I)=SPEC(I)
  IF(I.LE.160)NUMA(I)=0
2 CONTINUE
C
C**** SET UP EXTENDED MEMORY WITH INPUT DATA
C
CALL VMEM(IC,IE) ;how much ext. memory space is available?
CALL CHECK(IE)
IF(IC.LE.23)TYPE"EXT. MEMORY TOO SMALL, BLOCKS=" ,IC*4
CALL MAPDF(IC,I01,1,32,IE) ;map ext. memory with elements of 32
                                words long and place the window
                                array I01 at block 0
C
CALL CHECK(IE)
CALL ERDB(2,0,4,62,IC,IE) ;read into ext. mem. noise data

```

```

C      CALL ERDB(4,0,66,62,IC,IE) ;read into ext. mem. speech plus noise
C**** START PROCESS LOOP
C
C      IST=0
C      IST2=0
C      ICONT=1025
C
C**** CALCULATE AND STORE FFTS OF WINDOWED NOISE
C
C      3      IST=IST+1 ;increment counter
C      IF(IST.EQ.494) GO TO 16 ;check for processing complete
C      NST=IST+32 ;offset of 32 to first block of ext. mem.
C      CALL VFETCH(NUMA,NST,4) ;NUMA gets 128 words of noise at index NST
C      DO 4 M=1,1024 ;apply hamming window to NUMA and fill FOURN with
C          NUMA and the rest zeroes
C      IF(M.GE.129) FOURN(M)=CMPLX(0.0,0.0)
C      IF(M.GE.129) GO TO 4
C      FOURN(M)=CMPLX(NUMA(M),0.0)*HAMMING(M)
C      CONTINUE
C
C      4      CALL DFT5(FOURN,1024,0) ;calculate FFT of FOURN
C      IF(IST.GE.(NUMTF+1)) GO TO 5 ;check for storage filled-up
C      CALL WRBLK(1,((IST-1)*16),102,16,IE) ;store FFT
C      IF(IE.NE.1)TYPE"WRBLK1 ERROR"
C      IF(IST.EQ.NUMTF) GO TO 9 ;check for enough FFTs calculated
C      GO TO 3 ;go calculate and store more FFTs of noise
C      DO 6 N=0,(NUMTF-2) ;shift stored FFTs of noise
C          CALL RDBLK(1,((N+1)*16),104,16,IE)
C          IF(IE.NE.1)TYPE"RDBLK1 ERROR",IE
C          CALL WRBLK(1,(N*16),104,16,IE)
C          IF(IE.NE.1)TYPE"WRBLK2 ERROR",IE
C          CONTINUE
C      6      CALL WRBLK(1,((NUMTF-1)*16),102,16,IE) ;store new FFT in opened
C      7      ;space after shift
C      IF(IE.NE.1)TYPE"WRBLK3 ERROR",IE
C
C

```

```

C**** CALCULATE POWER SPECTRUM ESTIMATE OF NOISE
C
DO 8 I=1,1024 ;clear arrays involved in calculations
SPEC(I)=CMPLX(0.0,0.0)
FOURS(I)=SPEC(I)
8
9
DO 11 M=0,(NANAL-1) ;M represents reference frame
DO 10 N=-7,3 ;N is the shift q around M
NBLK=7+M+N ;calculation of location of FFT of frame
of shift (q+M)*32 samples
MBLK=7+M ;calculation of location of FFT of frame
of shift M*32 samples
CALL RDBLK(1,(NBLK*16),102,16,IE) ;read FFT into FOURN
IF(IE.NE.1)TYPE="RDBLK2 ERROR",IE
DO 10 I=1,1024 ;calculate sum of products of FFTs over
shift of N*32 for a value of M
A=I*-2*3.14159*N/32.0 ;calculate phase shift correction for shift
FOURS(I)=FOURS(I)+CONJG(CONJG(FOURN(I))*CMPLX(COS(A),SIN(A)))
CALL WRBLK(1,((NUMTF+NANAL+M)*16),104,16,IE) ;store sum over N
IF(IE.NE.1)TYPE="WRBLK4 ERROR",IE
CALL RDBLK(1,(MBLK*16),102,16,IE) ;read FFT of shift M*32
IF(IE.NE.1)TYPE="RDBLK3 ERROR",IE
DO 11 J=1,1024 ;calculate power spectrum from product of
sum over N and FFT of shift M*32
and add to power spectrum estimate
calculated for previous shift (M-1)*32
SPEC(J)=SPEC(J)+CONJG(FOURN(J))*FOURS(J)
FOURS(I)=CMPLX(0.0,0.0)
11
C
C**** CALCULATION OF CROSS SPECTRUM BETWEEN NOISE AND SPEECH PLUS NOISE
C
CALL SPCH(HAMMING,IST2,NANAL)
C
C**** CALCULATION OF WIENER FILTER AND FILTERING OF NOISE CHANNEL
C
CALL RDBLK(1,112,104,16,IE) ;read FFT of noise from frame M=0
CALL CHECK(IE)

```

```

DO 12 I=1,1024 ; calculate Weiner filter and multiply by FFT
      of noise (filtering)
      SPEC(I)=CONJG(FOURS(I))/1024*FOURN(I)/SPEC(I)
12      FOURS(I)=CMPLX(O.O,O.O)
C
C**** RECONSTRUCTION OF CORRECTION SIGNAL BY OVERLAP AND ADD METHOD
C
      CALL DFT5(SPEC,1024,1) ; inverse fourier transforming filtered noise
      IF(ICONT.NE.1025)CALL VFETCH(NUMA,ICONT-1,4) ; fetch previous samples
      of correction signal
DO 13 I=1,128 ; add present values to past values of correction signal
      with an overlap of 128-32=96 samples
      IF(ICONT.EQ.1025)NUMA(I+32)=0 ; if first time past samples =0
      NUMA(I)=REAL(SPEC(I))*46296+NUMA(I+32) ; 46296=scaling factor
13      CALL VSTASH(NUMA,ICONT,4) ; store new values of correction signal in
      extended memory at index ICONT
      CALL CHECK(IE)
      ICONT=ICONT+1 ; increment index of ext. memory
      IF(ICONT.LE.1273)GO TO 15 ; check for 31 blocks of signal generated
      IF(IST.GE.380)GO TO 15 ; precaution to insure the next sequence is
      executed once
C
C**** OUTPUT OF CORRECTION SIGNAL TO NSFILT
C
      CALL EWRB(5,0,128,31,IE) ; write to NSFILT first 31 blocks
      CALL CHECK(IE)
      ICONT=1026 ; reinitialize index
DO 14 I=0,1 ; clear extended memory for new values of signal
      CALL VSTASH(104,1025+I*128,128)
14      CALL VSTASH(NUMA,1025,4) ; store present values of signal
      GO TO 3 ; go and start process loop over again
15      CONTINUE
16      CALL EWRB(5,31,128,30,IE) ; write final 30 blocks to NSFILT
      CALL CHECK(IE)
      CALL RESET ; close channels
      END

```

```

C ***** THIS SUBROUTINE CALCULATES THE CROSS SPECTRUM BETWEEN SPEECH
C ***** PLUS NOISE AND NOISE CHANNELS
C
C ***** SET UP AND INITIALIZE ARRAYS
C
      SUBROUTINE SPCH(HAMMING,IST2,NANAL)
      DIMENSION NUMB(128)
      COMPLEX FOURN(1024),SPEC(1024),FOURS(1024),HAMMING(128)
      COMMON /BLK/IO1(1024)/BLK/IO2(4096),IO4(4096)
      EQUIVALENCE(FOURS(1),IO4(1))
      EQUIVALENCE(FOURN(1),IO2(1))
      DO 100 I=1,1024 ;clear arrays
      SPEC(I)=CMPLX(0.0,0.0)
      FOURN(I)=CMPLX(0.0,0.0)
      FOURS(I)=CMPLX(0.0,0.0)
      CONTINUE
100  NUMTF=NANAL+10 ;calculate # of FFTs used in power spectrum estimates
C
C ***** START CALCULATING AND STORING FFTS
C
      IST2=IST2+1 ;increment index
      NST=IST2+535 ;offset of 535 needed to read from ext. mem.
      CALL VFETCH(NUMB,NST,4) ;get 128 samples of speech plus noise at NST
      DO 68 M=1,1024 ;apply Hamming window to NUMB and fill FOURN with
      NUMB and the rest zeroes
      IF(M.GE.129)FOURN(M)=CMPLX(0.0,0.0)
      IF(M.GE.129)GO TO 68
      FOURN(M)=CMPLX(NUMB(M),0.0)*HAMMING(M)
      CONTINUE
68  CALL DFT5(FOURN,1024,0) ;calculate FFT of FOURN
      IF(IST2.GE.(NANAL+1)) GO TO 71 ;check for storage filled-up
      CALL WRBLK(1,((NUMTF+IST2-1)*16),IO2,16,IE) ;store FFT
      IF(IE.NE.1)TYPE="WRBLK2 ERROR2 ",IE
      IF(IST2.EQ.NANAL) GO TO 73 ;check for enough FFTs calculated

```

```

71      GO TO 63 ; if not go calculate FFT of another frame
      DO 72 N=0,(NANAL-2) ; shift stored FFTs on disk
          CALL RDBLK(1,((NUMTF+N+1)*16),IO4,16,IE)
          IF(IE.NE.1)TYPE"RDBLK4 ERROR2 ",IE
          CALL WRBLK(1,((NUMTF+N)*16),IO4,16,IE)
          IF(IE.NE.1)TYPE"RDBLK5 ERROR2 ",IE
72      CALL WRBLK(1,((NUMTF+NANAL-1)*16),IO2,16,IE) ; write FFT in opened
      ; space after shift
      IF(IE.NE.1)TYPE"RDBLK6 ERROR2 ",IE
      C
      C
      C**** CALCULATE CROSS POWER SPECTRUM ESTIMATE
      C
73      DO 107 M=0,(NANAL-1) ; M represents reference frame
          CALL RDBLK(1,((NUMTF+NANAL+M)*16),IO2,16,IE) ; read sum of products
          IF(IE.NE.1)TYPE"ERROR ON IO2 RDBLK ",IE
          CALL RDBLK(1,((M+NUMTF)*16),IO4,16,IE) ; read reference frame
          IF(IE.NE.1)TYPE"ERROR ON IO4 RDBLK",IE
          DO 110 J=1,1024 ; calculate power spectrum estimate from product of
              sum over N (calculated in main program) and
              FFT of frame M of speech plus noise and add to
              previous estimate
              SPEC(J)=SPEC(J)+CONJG(FOURS(J))*FOURN(J)
              IF(M.NE.(NANAL-1))GO TO 110 ; check for last estimation
              FOURN(J)=SPEC(J) ;FOURN gets estimate and interfaces to main
              ; program through labeled common
              SPEC(J)=CMPLX(0.0,0.0)
              CONTINUE
110      CONTINUE
107      RETURN
      END

```

[illegible]

```

DO 2 I=1,256 ; subtract with shift of 224 on speech plus noise
               because noise cancellation begins then
NADD(I)=NSPEECH(I+224)-NOISE(I)
CALL WRBLK(1,IST,NADD,1,IE) ; write noise cancelled speech
IF(IE.NE.1)TYPE="WRTBLK ERROR ",IE
IST=IST+1 ; increment counter
IF(IST.LE.NBLK) GO TO 100 ; is processing through
CALL DFILW("NSFILT",IE) ; delete correction signal file
CALL RESET
END

```

C 2

Noise Generation

The following source listing for NOISE explains the noise generation process.


```

ACCEPT "ENTER SEQUENCE LENGTH ",NSQ
ACCEPT "VARIANCE (0-2047) = ",VAR
CALL OPEN(12,"NUM3",3,IERR) ;open NUM3 to accept output noise
TYPE "ERROR CODE = ",IERR

C
C**** INITIALIZE ARRAY AND VARIABLES
C
A=16607.DO
DO 3 I=1,15872 ;clear XOUT
  XOUT(I)=0
3
PI=3.1415927
M=2147483647.DO ;Merseene prime
DELTA=1.0
IY=1.DO ;seed value for starting generator
CON=SQRT(2.0*PI*VAR) ;constant of gaussian distribution
CON1=-1.0*ALOG(.72E-76*CON)

C
C**** START GENERATOR LOOP
C
DO 10 I=1,NSQ
  IX=IY ;IX gets previous uniform deviate or seed (I=1)
  IX=A*IX ;A=16607
  IX=DMOD(IX,M) ;find remainder of IX/M (M=((2**32)-1))
  IY=IX ;IY is set up with uniform deviate
  IX=IX/M ;limit range of deviates to 0-1
  XX=IX ;double precision to single precision
  XNEW=0.0
  T=-1.0*SQRT(CON1*2.0*VAR) ;lowest value of gaussian deviates
  IT=IFIX(T)
  T=FLOAT(IT) ;get rid of fractional part
  ILIM=-2*IT ;the full range of gaussian deviates
  XOLD=0.72E-76 ;the first gaussian deviate
  DO 13 J=1,ILIM ;start integration but no farther the ILIM
    T=T+DELTA ;increment gaussian deviate
    XBEF=(1.0/CON)*EXP(-T*T/VAR/2.0) ;calculate probability of gaussian
    ;deviate
  13
10

```

```

C      XNEW=XNEW+((XOLD+XBEP)/2.0) ; calculate cumulative probability of
      gaussian deviate
      XOLD=XBEP ; old=new deviate
      IF(XNEW-XX)13,13,7 ; is cumulative probability of gaussian
      deviate equal to that of the uniform
C 13  CONTINUE
C      C**** OUTPUT NOISE
C 7   XOUT(1)=IFIX(T) ; output gets gaussian deviate
10  CONTINUE
      CALL WRBLK(12,0,XOUT,62,IERR)
      TYPE "IERR= ",IERR
      CALL RESET
      STOP
      END

```

Adding Noise to Speech

The following source listing for SPLUSN explains the process of adding noise to speech with a predefined signal to noise ratio.


```

ACCEPT" NOISE FILENAME ? "
READ(11,1) NOISFIL(1) ;read noise filename
ACCEPT" INPUT SIGNAL TO NOISE RATIO (DB) ",SNR
ACCEPT" DESIRE B WEIGHTING ON NOISE SPECTRA ? Y=1,N=0 ",NB
ACCEPT" DESIRE B WEIGHTING ON SPEECH SPECTRA ? Y=1,N=0 ",NSB
ACCEPT" OUTPUT FILENAME ? "
READ(11,1) OUTFILE(1) ;read output filename

```

```

C
C**** OPEN I/O FILES
C

```

```

CALL OPEN(2,NSPFIL,2,IE)
IF(IE.NE.1)TYPE"NSPCFIL OPEN ERROR ",IE
CALL OPEN(3,NOISFIL,2,IE)
IF(IE.NE.1)TYPE"NOISFIL OPEN ERROR ",IE
CALL OPEN(1,OUTFILE,2,IE)
IF(IE.NE.1)TYPE"OUTFILE OPEN ERROR ",IE

```

```

C
C**** CALCULATE ENERGY IN SPEECH AND NOISE
C

```

```

SPOWER=0.0
ENNOISE=0.0
KN=1
KS=1

```

```

DO 50 I=1,128 ;initialize B weighting array
K=I-1

```

```

50 B(I)=7160.0*7160.0*K*K/(K**4+4.90256E7*K*K+1.2544E12)

```

```

DO 110 I=0,87

```

```

CALL RDBLK(2,1,NSPEECH,1,IE) ;read a block of speech
CALL RDBLK(3,1,NOISE,1,IE) ;read a block of noise
DO 101 IM=1,256 ;initialize complex arrays with sampled data
FOURS(IM)=CMPLX(NOISE(IM),0.0)
FOURN(IM)=CMPLX(NSPEECH(IM),0.0)

```

```

101

```

```

CALL DFT5(FOURN,256,0) ;calculate DFT of speech
CALL DFT5(FOURS,256,0) ;calculate DFT of noise
DO 103 IF=1,128 ;calculate energy in speech and noise
IF(NB.EQ.1)KN=B(IF)

```

```

103 IF(NSB.EQ.1)KS=B(IF)
110 ENNOISE=ENNOISE+(CABS(FOURS(IF))**2)*KN/2.8836E6
      SPOWER=SPOWER+(CABS(FOURN(IF))**2)*KS/2.8836E6
      CONTINUE
      TYPE"SPEECH POWER= ",SPOWER
      CONSTANT=SQRT(SPOWER/ENNOISE/10**(SNR/10)) ; calculate gain applied to
                                                    noise before adding to
                                                    speech for SNR
      TYPE"CONSTANT= ",CONSTANT
      C
C**** ADD NOISE TO SPEECH FOR SNR AND OUTPUT SPEECH PLUS NOISE
      C
      DO 120 I=0,87
        CALL RDBLK(3,I,NOISE,1,IE) ;read block of noise
        CALL RDBLK(2,I,NSPEECH,1,IE) ;read block of speech
        DO 119 J=1,256 ;add noise to speech after gain is applied to noise
          NOUT(J)=NSPEECH(J)+IFIX(CONSTANT*FLOAT(NOISE(J)))
          CALL WRBLK(1,I,NOUT,1,IE)
        CONTINUE
      TYPE"NOISE POWER= ",ENNOISE
      END

```


APPENDIX B

The Use of the ILS System

The use of the ILS system to perform linear predictive analysis/synthesis of a speech file is done by executing the following steps.

1. 2ILSFIL - The system asks you for the name of CHOPS file, the name of ILS file to be created (WD#). # is a file number desired.
2. FIL WD# - Necessary to designate WD# as primary file.
3. INA - Initializes the LPC analysis requirements. The system will ask you for the values of these requirements.
4. API N1,N2 - Performs the LPC analysis from frame N1 to N2. N1 must be greater than or equal to 3. The API command takes speech information from the primary file WD# and stores the analysis parameters in a secondary file.
5. FIL U# - Unprotects file WD# so that the synthesis program can take the synthesized speech and store it back into the primary file WD#.

- 6. SNS
 - Performs synthesis of speech from parameters stored in secondary file.
- 7. AGC
 - Program which multiplies resulting speech file by a gain factor to prevent clipping when speech is outputted through the D/A converter.

Vita

Christopher Lee Batchelor was born on 22 December 1954 in Corona CA. He graduated from Havelock High School in Havelock NC in 1972. From August of 1973 to May of 1977, he attended North Carolina State University in Raleigh NC. There he received, as an Honor Graduate, the degree of Bachelor of Science in Electrical Engineering in May 1977. He then worked as Assistant Engineer for AAI, Inc. in Cockeysville MD with the duties of radar test set development. In August 1978 he went to work as Assistant Engineer for Babcock and Wilcox in Lynchburg VA with the duties of microprocessor based control system development. Next, in January 1980 he attended Officers' Training School and received a commission in the United States Air Force in June 1980. He then entered the School of Engineering of the Air Force Institute of Technology.

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BLOCK 20: Abstract (Cont'd):

preprocessing. Preliminary laboratory listenings verified that an improvement in quality was achieved with noise cancellation preprocessing. Although improvement in quality was achieved, more effort is required to make this implementation more efficient and improve the quality of speech produced.

21

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